

# Hyper-dimensional time-series data analysis with reservoir computing networks to predict plasma profiles in tokamak

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Collaboration: DIII-D Team

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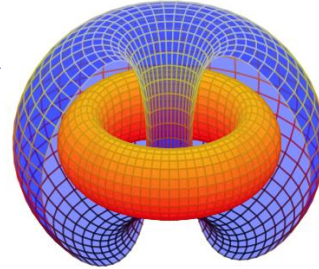
# Outline

- **What is tokamak?**
- **why data-driven models for plasma condition monitoring?**
- **What is Reservoir Computing Network?**
- **Real-time adaptive RCN models for plasma profile prediction**

# Plasma condition monitoring in Tokamaks

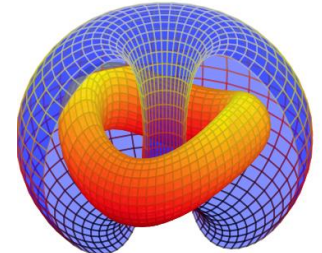


Stable Equilibrium

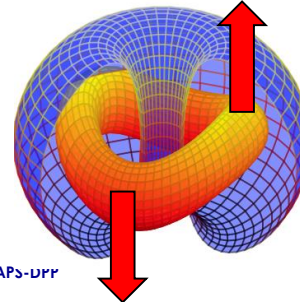


$\Delta t \sim 1s$

Unstable Equilibrium



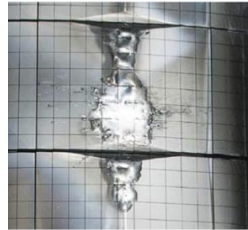
Unstable Mode Growth



$\Delta t \ll 1s$

## Example Wall Damage from JET

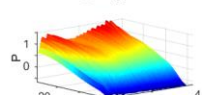
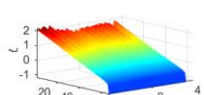
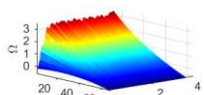
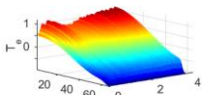
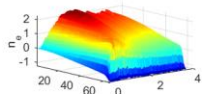
Matthews et al. Physica Scripta, T167, 2016



**ITER and future reactors  
need disruption avoidance solutions**

# Data-Driven Transport Model for DIII-D

Given state (and actuators), can we predict how plasma will evolve on transport timescales (~100-200ms)?

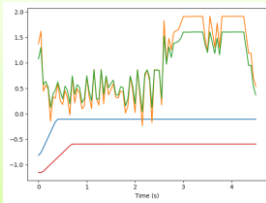


$x(t)$

- Electron temperature ( $T_e$ )
- Electron density ( $n_e$ )
- Safety factor ( $q$ )
- Rotation ( $\Omega$ )
- Pressure ( $P$ )

$u(t)$

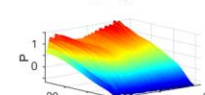
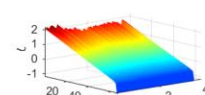
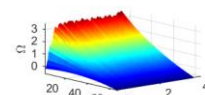
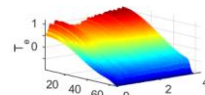
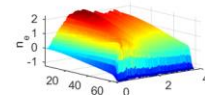
- $P_{injected}$
- $T_{injected}$
- $I_p$
- $\langle n_e \rangle_{target}$



$$x(t+\Delta t) = x(t) + f[x(t), u(t)]$$

$f[x(t), u(t)]$

- $T_e$
- $n_e$
- $q$
- $\Omega$
- $P$



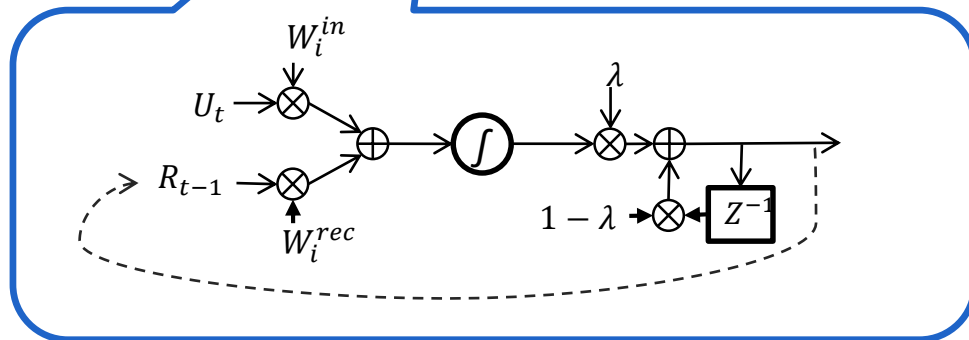
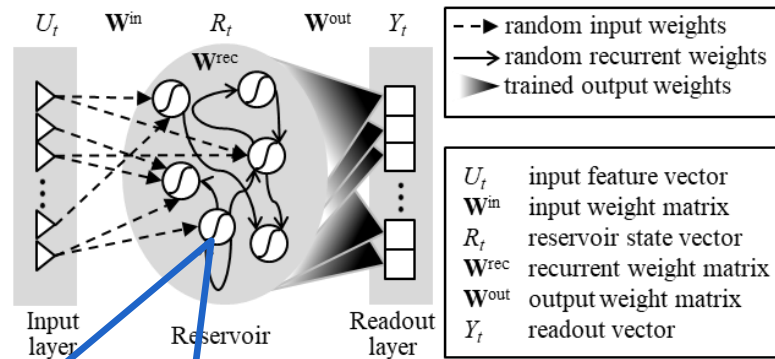
# What is Reservoir computing network\*?

\* Jaeger H. (2001) The "echo state" approach to analyzing and training recurrent neural networks. GMD Report 148, GMD - German National Research Institute for Computer Science

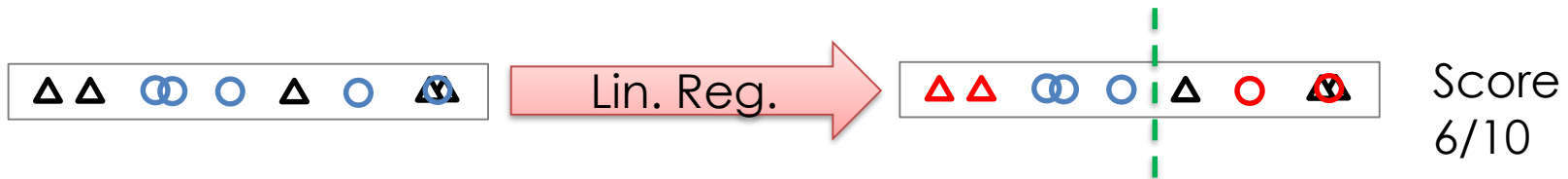
- **Reservoir computing network (RCN)**

- Recurrent neural network
- Randomly connected early layers.
- Only Last layer is trained using linear regression.
- Suitable for time-series data analysis
- **Much faster and easier training** procedure compared to DNNs.

The model can be **adapted** to the new data in **real-time!**  
Adaptation takes only **100ms** on a laptop CPU!



# How does RCN work?

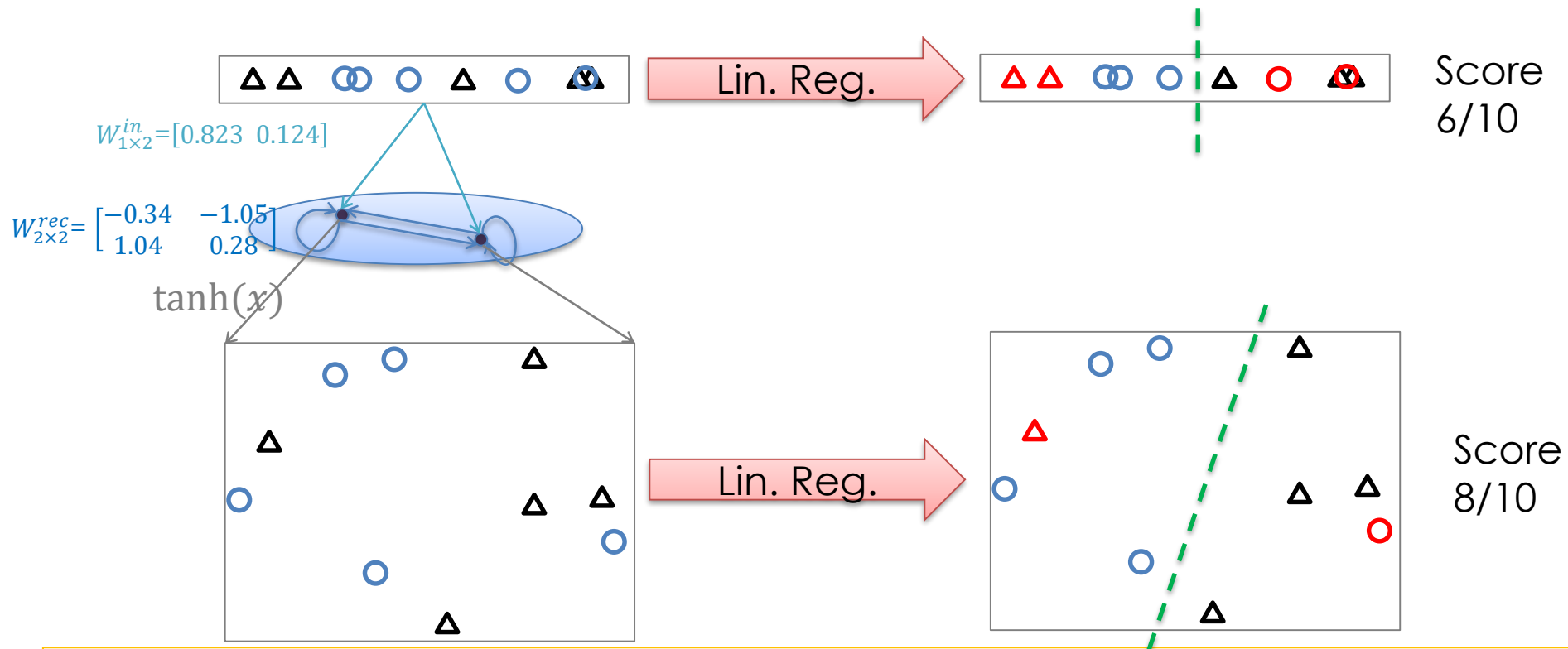


# How does RCN work?



$W^{in}$  and  $W^{rec}$  are set randomly

# How does RCN work?

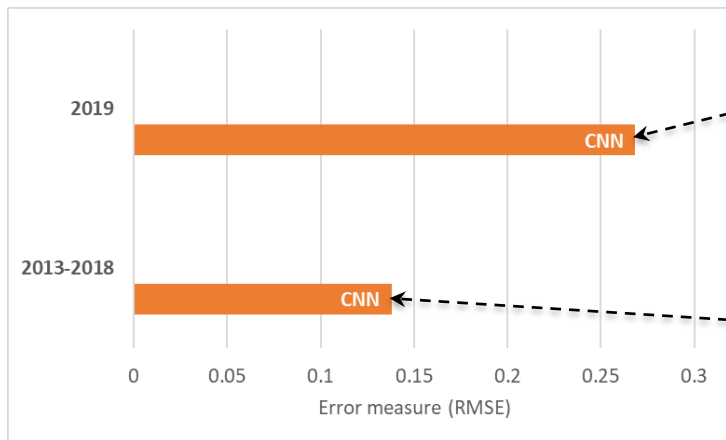


RCN randomly projects the inputs to a (much) larger non-linear feature space.



# Adaptive Data-driven Profile Prediction Models

	CNN/LSTM
Training	5h on GPU
Performance	SOTA



Train a on the data from **2013-2018**

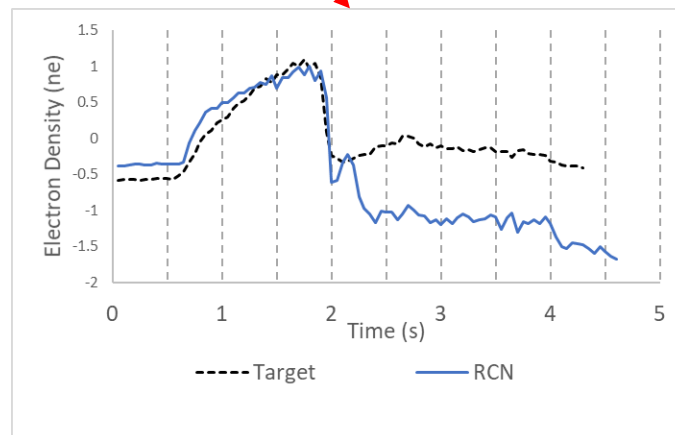
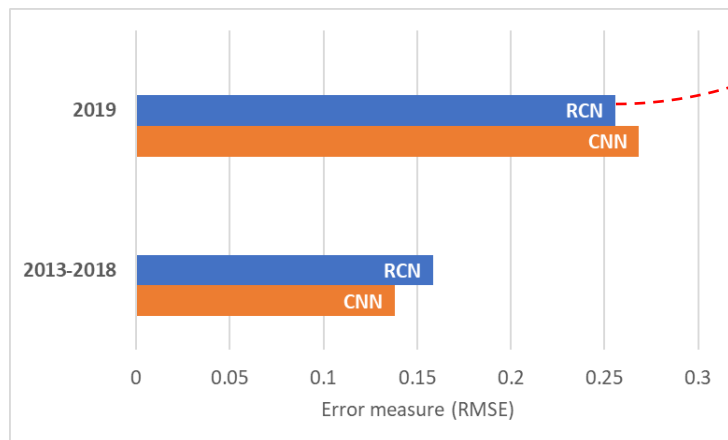
Predict the profiles for **2019**

Predict the profiles for **the same period**

# Adaptive Data-driven Profile Prediction Models

By means of Reservoir Computing Network

	CNN/LSTM	RCN
Training	5h on GPU	45s on CPU
Performance	SOTA	Close to DNN

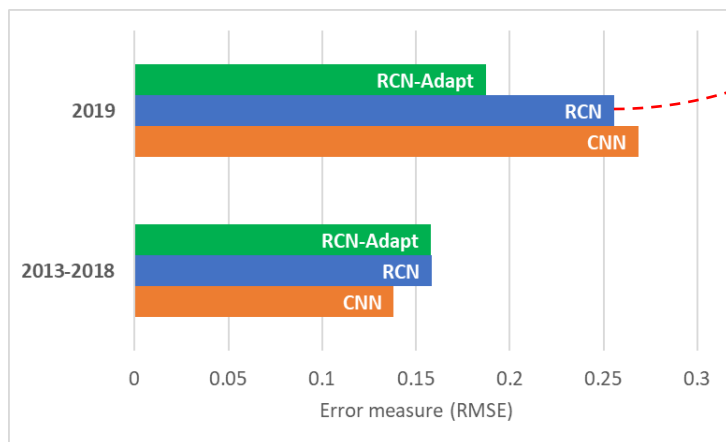


# Adaptive Data-driven Profile Prediction Models

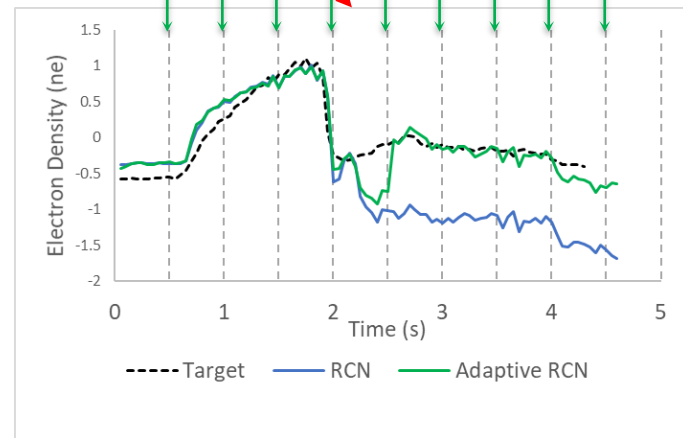
By means of Reservoir Computing Network \*

\* A. Jalalvand, J. Abbate, R. Conlin, G. Verdoolaege, E. Kolemen, "Real-Time and Adaptive Reservoir Computing with an Application to Profile Prediction in Fusion Plasma", IEEE Trans. on Neural Net. & Learning Systems, 2020.

	CNN/LSTM	RCN
Training	5h on GPU	45s on CPU
Performance	SOTA	Close to DNN
Adaptation	Difficult	Easy (100ms)



Real-time Adapting every 500ms



# Conclusion

- RCN performs similar to state-of-the-art DNN models in predicting plasma profiles.
- Easy and fast training procedure of RCN is an important asset for quick adaptation to the new environment.

Thank you!